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CONVERGENCE AND ASYMPTOTIC VARIANCE OF BOOTSTRAPPED FINITE-TIME RUIN PROBABILITIES WITH PARTLY SHIFTED RISK PROCESSES

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Convergence and asymptotic variance of bootstrapped finite-time ruin probabilities with partly shifted risk processes.

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Abstract. The classical risk model is considered and a sensitivity analysis of finite-time ruin probabilities is carried out. We prove the weak convergence of a sequence of empirical finite-time ruin probabilities. So-called partly shifted risk processes are introduced, and used to derive an explicit expression of the asymptotic variance of the considered estimator. This provides a clear representation of the influence function associated with finite time ruin probabilities, giving a useful tool to quantify estimation risk according to new regulations.

Keywords: Finite-time ruin probability, robustness, Solvency II, reliable ruin probability, asymptotic normality, influence function, partly shifted risk process, Estimation Risk Solvency Margin (ERSM).

AMS Classification: 15A15, 15A09, 15A23

1. Introduction

The surplus of an insurance company is classically modeled by the risk process $(R_t)_{t \geq 0}$ defined as follows : for $t \geq 0$,

$$R_t = u + ct - S_t,$$

where u is the non-negative amount of initial reserves and $c > 0$ is the premium income rate. The cumulated claim amount up to time t is described by the compound Poisson process

$$S_t = \sum_{i=1}^{N_t} W_i,$$

where the amounts of claims $W_i, i = 1, 2, \dots$ are non-negative independent, identically-distributed random variables, distributed as W , with the convention that $S_t = 0$ if $N_t = 0$. The number of claims N_t until $t \geq 0$ is modeled by an homogeneous Poisson process $(N_t)_{t \geq 0}$ of intensity λ . Claim amounts and arrival times are assumed to be independent.

We are interested in the robust estimation of finite-time ruin probabilities. Solvency regulations for insurance companies, called *Solvency II*, impose the control of a certain number of insolvency probabilities. The chosen risk measure to determine the Solvency Capital Requirements (SCR) is more likely to be a 99.5%, one-year Value at Risk than a continuous-time ruin probability. Nevertheless, reserving is expected to be quantified by a best estimate of liabilities, plus a so-called Market-Value Margin (MVM), which is determined by a cost-of-capital approach: this margin corresponds to the cost of maintaining the surplus above the SCR level during the whole period $[0, t]$, where t is typically 10 years. This corresponds to a continuous-time ruin problem in finite horizon. Let us denote by $\psi(u, t)$ the probability of ruin before time t with initial reserve u :

$$\psi(u, t) = \mathbb{P}[\exists s \in [0, t], \quad R_s < 0 \mid R_0 = u], \quad u \geq 0, t > 0,$$

and let

$$\varphi(u, t) = 1 - \psi(u, t)$$

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be the probability of non-ruin within time t with initial reserve u .

Algorithms to compute or approximate $\psi(u, t)$ have been proposed, among others, by (Asmussen, Avram and Usabel (2002)) by the means of Erlangization, by (Picard and Lefèvre(1997)) by the means of Appell polynomials, and by (Rullière and Loisel (2004)) by the means of a *Seal-type* argument based on the ballot Lemma. An important feature of Solvency II is that estimation risk should be controlled, particularly if internal models are used. However, most models use a calibrated model and an *a posteriori* proportional loading factor to take into account estimation risk. It would be of course much better to carry out a robustness analysis at the same time. This led us to define in an earlier paper (see (Loisel, Mazza and Rullière(2007))) reliable ruin probabilities as quantiles of empirical finite-time ruin probabilities, and the Estimation Risk Solvency Margin (ERSM) as the additional capital required to cover estimation risk:

Let $\psi^N(u, t)$ and $\varphi^N(u, t)$ respectively be the finite-time ruin and non-ruin probability with claim amounts drawn from the empirical distribution F_N associated with an i.i.d. sample of distribution F , where F is the c.d.f. of W and $N \geq 1$ is the size of the sample. Define the *reliable finite-time ruin probability* $\psi_{1-\varepsilon}^{N, \text{reliable}}(u, t)$ as the $(1 - \varepsilon)$ -quantile of the (random) bootstrapped finite-time ruin probability $\psi^N(u, t)$:

$$\psi_{1-\varepsilon}^{N, \text{reliable}}(u, t) = \inf_{s \geq 0} \left\{ \mathbb{P} \left[\psi^N(u, t) \geq s \right] \leq \varepsilon \right\}.$$

If u_η and $u_{\eta, \varepsilon}$ are respectively defined as the initial capital required to ensure that

$$\psi(u_\eta, t) \leq \eta$$

and

$$\psi_{1-\varepsilon}^{N, \text{reliable}}(u_{\eta, \varepsilon}, t) \leq \eta,$$

the Estimation Risk Solvency Capital $ERSM_{\eta, 1-\varepsilon}$ can be defined as the additional capital needed to take estimation risk into account:

$$ERSM_{\eta, 1-\varepsilon} = u_{\eta, 1-\varepsilon} - u_\eta.$$

In (Loisel, Mazza and Rullière(2007)), we have shown the convergence of

$$\sqrt{N} \left(\varphi(u, t) - \varphi^N(u, t) \right)$$

as N tends to $+\infty$ in distribution to a centered, Gaussian random variable only for $u = 0$. The proof was based on the symmetrical (in W_1, \dots, W_n) expression

$$\varphi(0, t) = \sum_{n \geq 1} P(N_t = n) E \left[(u + ct - (W_1 + \dots + W_n))_+ \right],$$

where x_+ denotes the positive part of a real number x . We also computed the asymptotic variance V_u of the limit of

$$\sqrt{N} \left(\varphi(u, t) - \varphi^N(u, t) \right),$$

and expressed this variance in terms of the variance of a random variable defined from the influence function of finite-time non ruin probability. Influence functions were introduced in the field of robust statistics to study the impact of data contamination on the estimated quantity (see (Huber(1981)) and (Hampel(1974))). For a functional \mathbf{T} of a distribution F , the influence function at point $x \in \mathbf{R}$ is defined as the limit (when it exists)

$$\text{IF}_x[\mathbf{T}] = \lim_{s \downarrow 0} \frac{\mathbf{T}(F^{(s, x)}) - \mathbf{T}(F)}{s},$$

where $F^{(s, x)}$ is defined for $x \in \mathbf{R}$ and $s > 0$ by

$$\text{for } u \in \mathbf{R}, \quad F^{(s, x)}(u) = s 1_{\{x \leq u\}} + (1 - s)F(u).$$

In the sequel, for each quantity related to the contaminated distribution $F^{(s,x)}$, we use the exponent $^{(s,x)}$. In a recent paper, (Marceau and Rioux(2001)) provided an algorithm to compute the influence function of the eventual probability of ruin. We obtained in (Loisel, Mazza and Rullière(2007)) that for all $u \geq 0$,

$$V_u = V_Y [\text{IF}_Y [\varphi(u, t)]] . \quad (1)$$

Nevertheless, it remains to be shown that the limit distribution of

$$\sqrt{N} \left(\varphi(u, t) - \varphi^N(u, t) \right)$$

is Gaussian, and besides formula (1) involves computation of influence functions described in (Loisel, Mazza and Rullière(2007)), which corresponds to heavy computation times and new convergence issues. In this paper, we prove the convergence in distribution of

$$\sqrt{N} \left(\varphi(u, t) - \varphi^N(u, t) \right)$$

toward a Gaussian random variable for all $u \geq 0$ by the means of U -statistics and so-called partly shifted risk processes. These processes are defined in section 2, in which finite-time ruin probabilities for partly shifted risk processes are computed as well. The expression of V_u in terms of ruin probabilities for modified risk processes derived in section 3.5 is of fundamental importance from a theoretical and operational point of view: it gives a probabilistic representation of V_u , which is used to prove the convergence of the empirical ruin probability for arbitrary $u \geq 0$. We also give elegant mathematical expressions for influence functions associated to finite time ruin probabilities. Finally, we provide efficient numerical methods for computing V_u .

2. Finite-time ruin probabilities for partly shifted risk processes

2.1. PARTLY SHIFTED RISK PROCESSES

Given $x \geq 0$, we define the x -partly shifted risk process as the stochastic process given by

$$R_t^x = u + ct - S_t^x,$$

where

$$S_t^x = S_t + x1_{\{U \leq t\}},$$

and U is a certain positive random variable. After this random delay U , the sample path of R_t^x is shifted x units downwards. It corresponds to add a jump of size x at a random instant U .

The process $(R_t^x)_{t \geq 0}$ has no longer stationary and independent increments. Nevertheless, we will show how to adapt results of risk theory to these partly shifted risk processes. This is important since, as we will see, finite-time ruin probabilities for partly shifted processes are directly involved in computations of sensitivities, influence functions and asymptotic variance of finite-time ruin probabilities for classical risk processes.

2.2. FINITE-TIME RUIN PROBABILITY FOR PARTLY SHIFTED RISK PROCESSES STARTING FROM ZERO

Let us consider the case where claim amounts are integer-valued.

PROPOSITION 1. *For $c > 0$, $x > 0$ and $n \in \mathbf{N} \setminus \{0\}$, we have*

$$\mathbf{P} \left[S_t^x < ct, \forall t < \frac{n}{c} \right] = \mathbf{P} \left[S_{i/c}^x < i, i = 1..n \right] .$$

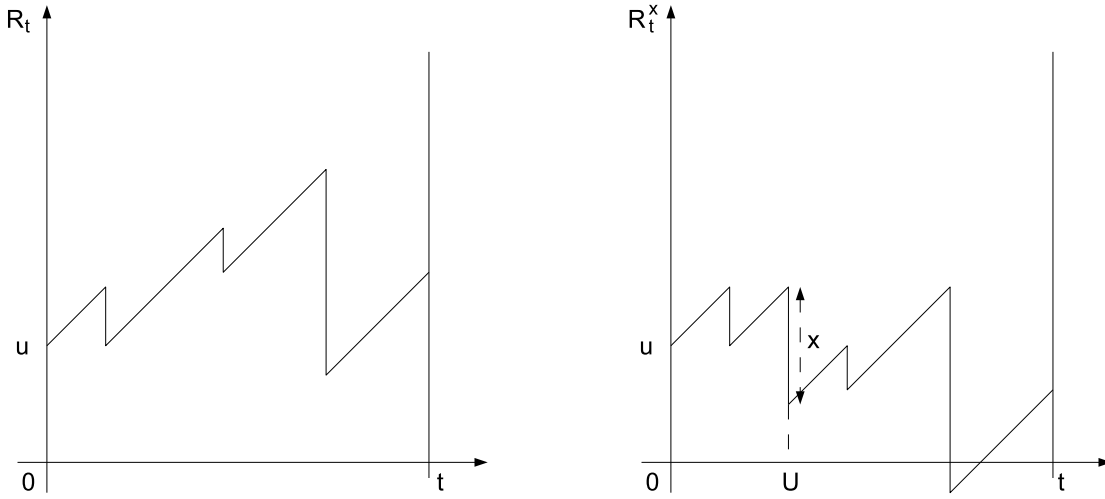


Figure 2. A corresponding sample path of the x -partly shifted risk process.

This is a direct adaptation of results of (Rullière and Loisel (2004)). It thus suffices to study the cumulated claim amount process at a finite set of inventory dates.

To this end, consider the random variables defined by $Y_i = S_{i/c}$, $Y_i^x = S_{i/c}^x$, and $\bar{U} = [U + 1]$, where $[x]$ denotes the floor of real number x . The finite-time ruin probability for partly shifted risk processes starting from 0 is obtained by an adaptation of the well known ballot Theorem.

LEMMA 1 (Ballot Theorem (see Takács (1962,a,b))). *For $n \geq 1$, let Z_i , $i = 1..n$ be a process with exchangeable increments. Then, we have*

$$\mathbb{P}[Z_i < i, i = 1..n, Z_n = j] = \frac{n-j}{n} \mathbb{P}[Z_n = j] \quad 0 \leq j \leq n.$$

The ballot Theorem applies to the family of random variables Y_i , $i = 1, \dots, n$, which has i.i.d. increments, but also to the Y_i^x , $i = 1, \dots, n$.

PROPOSITION 2 (Ballot theorem for partly shifted risk processes). *Let W_1, \dots, W_n be i.i.d., integer-valued random variables. Consider the partial sum process*

$$Y_i = \sum_{j=0}^i W_j,$$

and define

$$Y_i^x = Y_i + x 1_{\{\bar{U} \leq i\}},$$

where \bar{U} uniformly distributed on the finite set $\{1/n, 2/n, \dots, 1\}$. Then,

$$\mathbb{P}[Y_i^x < i, i = 1..n, Y_n^x = j] = \frac{n-j}{n} \mathbb{P}[Y_n^x = j] \quad 0 \leq j \leq n.$$

Proof: Takács's result is true for exchangeable random variables (and even for cyclically exchangeable random variables). It can easily be shown that the Y_i^x , $1 \leq i \leq n$ are exchangeable. To this end, consider the distribution function

$$F_{Y_{\sigma(1)}^x, \dots, Y_{\sigma(n)}^x}$$

associated with the random vector

$$(Y_{\sigma(1)}^x, \dots, Y_{\sigma(n)}^x),$$

which is also the distribution function of the random vector

$$(Y_1^x, \dots, Y_n^x),$$

which is denoted by

$$F_{Y_1^x, \dots, Y_n^x},$$

$\forall y_1, \dots, y_n$. Then

$$F_{Y_{\sigma(1)}^x, \dots, Y_{\sigma(n)}^x}(y_1, \dots, y_n) = \frac{1}{n} \sum_{i=1}^n F_{Y_{\sigma(1)}^x, \dots, Y_{\sigma(n)}^x}(y_1, \dots, y_{i-1}, y_i - x, y_{i+1}, \dots, y_n) \quad (2)$$

$$= \frac{1}{n} \sum_{i=1}^n F_{Y_1^x, \dots, Y_n^x}(y_1, \dots, y_{i-1}, y_i - x, y_{i+1}, \dots, y_n) \quad (3)$$

$$= F_{Y_1^x, \dots, Y_n^x}(y_1, \dots, y_n). \quad (4)$$

We used the fact that the random variables Y_i , $1 \leq i \leq n$, are i.i.d. to get (3) from (2), and the fact that U is uniformly distributed on $[0, t]$, and independent of the Y_i 's to write (2), and to get (4) from (3). \square

REMARK 1. *The previous result remains valid if U is uniformly distributed on $\{1, \dots, n_{\max}\}$, with $n_{\max} \geq n$. To prove this, distinguish two cases: given that $U \leq n$, proposition 2 applies, and given that $U > n$, the classical ballot lemma applies.*

Propositions 1 and 2 directly enable us to obtain the finite-time ruin probability for partly shifted risk processes starting from 0:

THEOREM 1. *The finite-time ruin probability for partly shifted risk processes starting from 0 is given by*

$$\mathbb{P}[R_s^x \geq 0 \forall s < t \mid R_0^x = 0] = \mathbb{E} \left[\frac{(ct - x - S_t)_+}{ct} \right],$$

where $R_s^x = R_s - x1_{\{U \leq s\}}$, and U is uniformly distributed on $[0, t]$.

2.3. FINITE-TIME RUIN PROBABILITY FOR PARTLY SHIFTED RISK PROCESSES STARTING FROM $u \geq 0$

Conditionally on the last *continuous* passage of R_t at 0, the process is located under the barrier $u + ct$ at time $t = n/c$ if there is no ruin, or if the last visit of the process at 0 occurred at time i/c . Let $T_u(x)$ be the first instant of ruin associated with the modified process, when $R_0 = u$. Then we have (see (Rullière and Loisel (2004)))

$$\mathbb{P} \left[S_{n/c}^x < u + n \right] = \mathbb{P} \left[T_u^x > \frac{n}{c} \right] + \Delta,$$

where

$$\Delta = \sum_{i=1}^{n-1} \sum_{j=1}^{n-i} \mathbb{P} \left[S_{i/c}^x = u + i \cap S_{i+k/c}^x - S_{i/c}^x < i, k = 1, \dots, n-i \cap S_{n/c}^x - S_{i/c}^x = j \right].$$

If \bar{U} is uniformly distributed on $1, \dots, n$, conditioning on \bar{U} , we can consider every element of the second sum and get the following result.

THEOREM 2 (modified ruin probability when $R_0 = u$). *The ruin probability associated to the modified process is given by*

$$\mathbb{P} \left[T_u^x > \frac{n}{c} \right] = \mathbb{P} \left[x + S_{n/c} < u + n \right] - \Delta_1 - \Delta_2,$$

where

$$\Delta_1 = \sum_{i=1}^{n-1} \frac{i}{n} \mathbb{P} \left[x + S_{i/c} = u + i \right] \sum_{j=1}^{n-i} \mathbb{P} \left[S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i},$$

and

$$\Delta_2 = \sum_{i=1}^{n-1} \frac{n-i}{n} \mathbb{P} \left[S_{i/c} = u + i \right] \sum_{j=1}^{n-i} \mathbb{P} \left[x + S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i}.$$

3. Sensitivity analysis

3.1. THE RUIN PROBABILITY AS A FUNCTION OF THE ADDITIONAL CLAIM

In the sequel, we consider the probability of non-ruin before $t = n/c$, when starting with an integer valued initial reserve. Conditioning on the last passage time i/c of R_t at 0, the probability that the reserve is some integer j at time n/c , $1 \leq j \leq u + n$, is given by (see (Loisel, Mazza and Rullière(2007)))

$$\begin{aligned} \mathbb{P} \left[R_{n/c} = j \right] &= \mathbb{P} \left[T_u > n/c \text{ and } R_{n/c} = j \right] \\ &+ \sum_{i=1}^{n-j} \mathbb{P} \left[S_{i/c} = u + i \right] \mathbb{P} \left[S_{(n-i)/c} = j \text{ and } S_{\nu/c} < \nu \forall \nu, 1 \leq \nu \leq n-i \right], \end{aligned}$$

where we assume that the sum vanishes when $j \geq n$. Using classical results of (Takács(1962a)), one obtains that the first term is given by

$$\mathbb{P} \left[S_{n/c} = u + n - j \right] - 1_{\{j \leq n\}} \sum_{i=1}^{n-j} \mathbb{P} \left[S_{i/c} = u + i \right] \mathbb{P} \left[S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i}.$$

For $j = 0$, we have $\mathbb{P} \left[T_u > n/c \text{ and } R_{n/c} = 0 \right] = \mathbb{P} \left[S_{n/c} = u + n \right] - \mathbb{P} \left[S_{n/c} = u + n \right] = 0$. Conditionally on $\{N_{n/c} = k\}$, defining

$$\varphi_{k,j}(u, n) = \mathbb{P} \left[T_u > n/c \text{ and } R_{n/c} = j \mid N_{n/c} = k \right],$$

and $\varphi_k(u, n) = \mathbb{P} \left[T_u > n/c \mid N_{n/c} = k \right]$, one obtains that

$$\begin{aligned} \varphi_{k,j}(u, n) &= \mathbb{P} \left[W^{*k} = u + n - j \right] \\ &- \sum_{i=1}^{n-j} \sum_{n_0=0}^k \alpha_{i,k}(n_0) \mathbb{P} \left[W^{*n_0} = u + i \right] \mathbb{P} \left[W^{*(k-n_0)} = j \right] \frac{n-i-j}{n-i}, \end{aligned}$$

where $\alpha_{i,k}(n_0) = \mathbb{P} \left[N_{i/c} = n_0 \right] \mathbb{P} \left[N_{(n-i)/c} = k - n_0 \right] / \mathbb{P} \left[N_{n/c} = k \right]$, *id est*

$$\alpha_{i,k}(n_0) = \binom{n_0}{k} i^{n_0} (n-i)^{k-n_0} / n^k.$$

So we have $\varphi_k(u, n) = \sum_{j=0}^{u+n} \varphi_{k,j}(u, n)$ and $\varphi(u, n) = \sum_{k=0}^{u+n} \mathbb{P} \left[N_{n/c} = k \right] \varphi_k(u, n)$. In fact, every claim amount W takes here a positive integer value (one can assume that $W \neq 0$ by modifying λ).

3.2. INFERENCE ON RUIN PROBABILITIES

The probability of ruin starting with initial reserve $u \geq 0$ is here expressed as a U-statistics. The basic information we use is exposed in (Hoeffding(1948)) and (Von Mises(1947)).

Let σ denote a permutation of $[N] = \{1, \dots, N\}$. Assuming that the observed claims belong to the finite set $\{w_1, \dots, w_N\}$, $\varphi_{k,j}(u, n)$ can be estimated by using the statistics $\hat{\varphi}_{k,j}(u, n)$. Our estimator has the form

$$\hat{\varphi}_{k,j}(u, n) = \frac{1}{N!} \sum_{[\sigma]} \left[\tilde{I}_1^k(u + n - j) - \sum_{i=1}^{n-j} \sum_{n_0=0}^k \alpha_{i,k}(n_0) \tilde{I}_1^{n_0}(u + i) \tilde{I}_{n_0+1}^k(j) \frac{n - i - j}{n - i} \right],$$

where

$$\tilde{I}_{i_0}^{i_1}(j) = 1_{\{\sum_{i=i_0}^{i_1} w_{\sigma(i)} = j\}},$$

and where the sum $\sum_{[\sigma]}$ is performed over all the permutations of $[N]$. Summing over j , one gets a U-statistics $\varphi_k(n)$, which estimates $\varphi_k(n)$.

Our next task consists in proving that $\hat{\varphi}_{k,j}(u, n)$ corresponds to the probability one gets when using the empirical distribution associated to the observed claims as our claim distribution. We use the following formula

$$\hat{\varphi}_{k,j}(u, n) = \frac{1}{N^k} \sum_{i_1, \dots, i_k \in \{1, \dots, N\}} \left[I_1^k(u + n - j) - \sum_{i=1}^{n-j} \sum_{n_0=0}^k \alpha_{i,k}(n_0) I_1^{n_0}(u + i) I_{n_0+1}^k(j) \frac{n - i - j}{n - i} \right],$$

where $I_x^y(j) = 1_{\{w_{i_x} + \dots + w_{i_y} = j\}}$, $x, y \in \mathbf{N}$, $x \leq y$. The indicator functions $I_x^y(j)$ are defined for each multi-index $\vec{i} = (i_1, \dots, i_k)$ (omitted here). Given u and n , set

$$\hat{\varphi}_{k,j}(u, n) = \frac{1}{N^k} \sum_{i_1, \dots, i_k \in \{1, \dots, N\}} \Phi_{k,j}(i_1, \dots, i_k),$$

where

$$\Phi_{k,j}(i_1, \dots, i_k) = \Phi_{k,j}^{(1)}(i_1, \dots, i_k) - \Phi_{k,j}^{(2)}(i_1, \dots, i_k),$$

and

$$\begin{aligned} \Phi_{k,j}^{(1)}(i_1, \dots, i_k) &= I_1^k(u + n - j), \\ \Phi_{k,j}^{(2)}(i_1, \dots, i_k) &= \sum_{i=1}^{n-j} \sum_{n_0=0}^k \alpha_{i,k}(n_0) I_1^{n_0}(u + i) I_{n_0+1}^k(j) \frac{n - i - j}{n - i}. \end{aligned}$$

A basic result of (Hoeffding(1948)) holds when $\Phi_{k,j}(i_1, \dots, i_k)$ is symmetric as a function of i_1, \dots, i_k . We shall study questions of symmetry in the next Section.

3.3. ON THE SYMMETRY OF $\hat{\varphi}_{k,j}(u, n)$

First, $\Phi_{k,j}^{(1)}(i_1, \dots, i_k) = I_1^k(u + n - j)$ is clearly symmetric as a function of i_1, \dots, i_k . Next, one studies the symmetry of $\Phi_{k,j}^{(2)}(i_1, \dots, i_k)$. A transposition that permutes i_1 and i_ν yields an element of the form $I_1^{n_0}(u + i) I_{n_0+1}^k(j)$. This element is modified into

$$I_1^{n_0}(u + i + 1_{\{\nu > n_0\}}(w_{i_1} - w_{i_\nu})) I_{n_0+1}^k(j - 1_{\{\nu > n_0\}}(w_{i_1} - w_{i_\nu})),$$

which is difficult to manage. We proceed differently and observe that $\alpha_{i,k}(n_0) = \mathbb{P}[B_{k,i/n} = n_0]$, where $B_{k,i/n}$ is the binomial coefficient of parameters k and i/n . Let σ_k denote all the possible permutations of the set $\{i_1, \dots, i_k\}$. Then one obtains that

$$\frac{1}{k!} \sum_{\sigma_k} \sum_{n_0=0}^k \alpha_{i,k}(n_0) I_1^{n_0}(u + i) I_{n_0+1}^k(j) = E_{\delta_1, \dots, \delta_k} [J_k(u + i) \bar{J}(j)],$$

with

$$J_k(u+i) = 1_{\{\delta_1 w_{i_1} + \dots + \delta_k w_{i_k} = u+i\}},$$

and

$$\bar{J}_k(j) = 1_{\{\bar{\delta}_1 w_{i_1} + \dots + \bar{\delta}_k w_{i_k} = j\}},$$

where $\bar{\delta}_\nu = 1 - \delta_\nu$, and where the random sequence δ_ν is i.i.d., with Bernoulli distribution with parameter i/n , $\nu \leq k$.

We can show that a transposition of two elements i_1 and i_ν has the same effect than a transposition of δ_1 and δ_ν . The random variables δ_ν are i.i.d., so that the mean is invariant under any permutation of the variables δ_ν , that is

$$E_{\delta_1, \dots, \delta_k} [J_k(u+i) \bar{J}(j)]$$

is symmetric as a function of i_1, \dots, i_k .

The point here is that $\sum_{n_0=0}^k \alpha_{i,k}(n_0) I_1^{n_0}(u+i) I_{n_0+1}^k(j)$ is not symmetric. We will use the following Lemma:

LEMMA 2.

$$N^k \hat{\varphi}_{k,j}(u, n) = \binom{N}{k} U + \sum^* \Phi_{k,j}(i_1, \dots, i_k),$$

where the sum \sum^* is taken over all k -uplets for which at least one pair of indices is such that $i_{\nu_0} = i_{\nu_1}$ ($\nu_0 \neq \nu_1$), and where U is a U -statistic.

Proof: Let $\Theta_k = \{i_1, \dots, i_k\}$ be a subset of $[N]$ of k distinct elements $i_1 < \dots < i_k$. Then one can write

$$\begin{aligned} \sum_{i_1, \dots, i_k \in \{1, \dots, N\}} \Phi_{k,j}^{(2)}(i_1, \dots, i_k) &= \sum_{\Theta_k \subset [N]} \sum_{\sigma_k} \sum_{i=1}^{n-j} \sum_{n_0=0}^k \\ \alpha_{i,k}(n_0) I_1^{n_0}(u+i) I_{n_0+1}^k(j) &\frac{n-i-j}{n-i} + \sum^* \dots \end{aligned}$$

As state previously, the expectation $E_{\delta_1, \dots, \delta_k} [J_k(u+i) \bar{J}(j)]$ is independent of the choice of the subset $\Theta_k \subset [N]$. one obtains that

$$\sum_{i_1, \dots, i_k \in \{1, \dots, N\}} \Phi_{k,j}^{(2)}(i_1, \dots, i_k) = \sum_{\Theta_k \subset [N]} \sum_{i=1}^{n-j} k! E_{\delta_1, \dots, \delta_k} [J_k(u+i) \bar{J}(j)] \frac{n-i-j}{n-i} + \sum^* \dots$$

where

$$\begin{aligned} U &= U^{(1)} - U^{(2)} \\ U^{(1)} &= \frac{1}{\binom{N}{k}} \sum_{\Theta_k \subset [N]} I_1^k(u+n-j), \end{aligned}$$

and

$$U^{(2)} = \frac{1}{\binom{N}{k}} \sum_{\Theta_k \subset [N]} \sum_{i=1}^{n-j} k! E_{\delta_1, \dots, \delta_k} [J_k(u+i) \bar{J}(j)] \frac{n-i-j}{n-i},$$

proving the result. □

3.4. Hoeffding's results

The variance of $\hat{\varphi}_{k,j}(u, n)$ is bounded and may be expressed as a function of a U -statistic. We can thus apply a powerful Theorem from Hoeffding (Theorem 7.3 on page 308 of (Hoeffding(1948))) to get the following Theorem.

THEOREM 3 (Asymptotic normality of the empirical ruin probability). *Let*

$$\hat{\varphi}_k(u, n) = \sum_{j=0}^{u+n} \hat{\varphi}_{k,j}(u, n),$$

and set

$$\Delta_k = \sqrt{N} (\hat{\varphi}_k(u, n) - \varphi_k(u, n)).$$

The random vector $(\Delta_1, \dots, \Delta_{u+n})$ is asymptotically centered normal of variance

$$\Gamma = \{\gamma \delta \zeta_1^{(\gamma, \delta)}\}_{\gamma, \delta \in \{1, \dots, u+n\}}.$$

The variance $\zeta_1^{(\gamma, \delta)}$ is not very explicit. We will discuss its various properties in the next Section.

REMARK 2. *We can assume that the number of claims is smaller than $u + n$, since, if not, ruin occurs with probability one: $W \geq 1$ and therefore $\Delta_k = 0$.*

The variance of $\sqrt{N} (\hat{\varphi}(u, n) - \varphi(u, n))$ is given by

$$V_u = \sum_{\gamma=0}^{u+n} \sum_{\delta=0}^{u+n} \gamma \delta \mathbb{P} [N_{n/c} = \gamma] \mathbb{P} [N_{n/c} = \delta] \zeta_1^{(\gamma, \delta)}.$$

If asymptotic normality holds, the limiting variance is given by formula (7) of (Loisel, Mazza and Rullière(2007)), and we have $V_u = V_Y [IF_Y(\varphi(u, t))]$, where IF_Y is the related influence function (for more details, see (Hampel(1974)), (Hampel et al.(1981)), or (Huber(1981))). Let V_Y denote the variance of the random variable Y . We will give equivalent expressions for these variances using the shifted ruin process of Section 2.

3.5. ALTERNATIVE FORMULAS FOR V_u

Given that among k claims, one of them is given by Y , one has $\varphi_{k,j}^Y(u, n) = L_{k,j}(Y) - R_{k,j}(Y)$. We omit Y when there is no ambiguity; furthermore, the letters L et R are used for *Left* and *Right*. Note that

$$L_{k,j} = \mathbb{P} [Y + W^{*k} = u + n - j].$$

Since we do not know a priori whether the claim Y occurred during the first n_0 claims, $R_{k,j}$ is obtained by conditioning on δ , the Bernoulli random variable of parameter i/n , which indicates if the claim occurred. When $j = 0$, one has $L_{k,0} = R_{k,0} = \mathbb{P} [Y + W^{*k} = u + n]$. When $j \geq 1$, conditioning on δ , one obtains that $R_{k,j}$ is given by

$$\begin{aligned} & \sum_{i=1}^{n-j} \sum_{n_0=0}^k \frac{i}{n} \mathbb{P} [B_{k-1,i/n} = n_0 - 1] \mathbb{P} [Y + W^{*n_0} = u + i] \mathbb{P} [W^{*k-n_0} = j] \frac{n-i-j}{n-i} \\ & + \frac{n-i}{n} \mathbb{P} [B_{k-1,i/n} = n_0] \mathbb{P} [W^{*n_0} = u + i] \mathbb{P} [Y + W^{*k-n_0-1} = j] \frac{n-i-j}{n-i}. \end{aligned}$$

Set

$$\begin{aligned} L_k(Y) &= \sum_{j=0}^{u+n} L_{k,j}(Y), \\ R_k(Y) &= \sum_{j=0}^{u+n} R_{k,j}(Y), \end{aligned}$$

$\bar{L}_k = E_Y [L_k(Y)]$ and $\bar{R}_k = E_Y [R_k(Y)]$. Then, following ((Hoeffding(1948)), Section 6), one gets that the covariance term $\zeta_1^{(k,l)}$ can be written as

$$\zeta_1^{(k,l)} = E_Y [(L_k - R_k) - (\bar{L}_k - \bar{R}_k)) ((L_l - R_l) - (\bar{L}_l - \bar{R}_l))] .$$

Hence, using the fact that both \bar{L} and \bar{R} are expectations over Y , we have

$$\zeta_1^{(k,l)} = E_Y [(L_k - R_k)(L_l - R_l)] - (\bar{L}_k - \bar{R}_k)(\bar{L}_l - \bar{R}_l).$$

In the sequel, if no specific indication is given, we consider that all probabilities, expectations and variances are taken given Y on the remaining random variables W_1, W_2, \dots

The variance V_u is given by

$$V_u = \sum_{k=1}^{\infty} \sum_{l=1}^{\infty} kl P [N_{n/c} = k] P [N_{n/c} = l] \zeta_1^{(k,l)},$$

and it is easy to show that in fact

$$V_u = V_Y \left[\sum_{k=1}^{+\infty} k P [N_{n/c} = k] (L_k(Y) - R_k(Y)) \right].$$

Using the fact that

$$k P [N_{n/c} = k] = \frac{\lambda n}{c} P [N_{n/c} = k - 1], \quad k \geq 1,$$

one gets that V_u is the variance of the random variable

$$\frac{\lambda n}{c} \sum_{k=0}^{+\infty} P [N_{n/c} = k - 1] (L_k(Y) - R_k(Y)).$$

The term on the left is given by

$$\begin{aligned} \sum_{k=0}^{+\infty} P [N_{n/c} = k - 1] L_{k,j}(Y) &= P [Y + S_{n/c} = u + n - j] \\ \sum_{k=0}^{+\infty} P [N_{n/c} = k - 1] L_k(Y) &= P [Y + S_{n/c} \leq u + n] \end{aligned}$$

On the other hand, the term on the right is obtained by using the identity

$$P [N_{n/c} = k] P [B_{k,i/n} = n_0] = P [N_{i/c} = n_0] P [N_{(n-i)/c} = k - n_0],$$

since we can write

$$\sum_{k=0}^{+\infty} P [N_{n/c} = k - 1] R_{k,j}(Y) = \sum_{i=1}^{n-j} \frac{n-i-j}{n-i} (S_{j,1} + S_{j,2}),$$

where $S_{j,1}$ and $S_{j,2}$ are given by

$$\sum_{k \in \mathbf{N}} \sum_{n_0 \leq k} \frac{i}{n} P [N_{i/c} = n_0 - 1] P [N_{(n-i)/c} = k - n_0] P [W^{*n_0-1} + Y = u + i] P [W^{*k-n_0} = j],$$

and

$$\sum_{k \in \mathbf{N}} \sum_{n_0 \leq k} \frac{n-i}{n} P [N_{i/c} = n_0] P [N_{(n-i)/c} = k - 1 - n_0] P [W^{*n_0} = u + i] P [W^{*k-n_0-1} + Y = j].$$

Hence, we have

$$\begin{aligned} S_{j,1} &= \frac{i}{n} \mathbb{P} \left[S_{i/c} + Y = u + i \right] \mathbb{P} \left[S_{(n-i)/c} = j \right], \\ S_{j,2} &= \frac{n-i}{n} \mathbb{P} \left[S_{i/c} = u + i \right] \mathbb{P} \left[S_{(n-i)/c} + Y = j \right]. \end{aligned}$$

Finally, one gets the following result.

THEOREM 4 (Asymptotic variance V_u). *Let Y be distributed according to the claim size distribution. V_u is the variance over Y of the following function of Y*

$$\begin{aligned} &\frac{\lambda n}{c} \mathbb{P} \left[Y + S_{n/c} < u + n \right] \\ &- \frac{\lambda n}{c} \sum_{i=1}^{n-1} \frac{i}{n} \mathbb{P} \left[S_{i/c} + Y = u + i \right] \sum_{j=1}^{n-i} \mathbb{P} \left[S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i} \\ &- \frac{\lambda n}{c} \sum_{i=1}^{n-1} \frac{n-i}{n} \mathbb{P} \left[S_{i/c} = u + i \right] \sum_{j=1}^{n-i} \mathbb{P} \left[S_{(n-i)/c} + Y = j \right] \frac{n-i-j}{n-i} \end{aligned}$$

We therefore find a mathematical expression which is similar to the ruin probability. This can be seen intuitively quite clearly since, apart from a factor $\lambda n/c$, the random variable of interest corresponds to the ruin probability associated to the process obtained by adding an additional claim Y . Looking at the various terms of the above expression, Y is added to $S_{i/c}$ with probability i/n , or to $S_{(n-i)/c}$ with probability $n-i/n$. In the special case where $Y = 0$, we recover the ruin probability obtained by summing over j :

$$\begin{aligned} &\mathbb{P} \left[T_u > n/c \text{ and } R_{n/c} = j \right] \\ &= \mathbb{P} \left[S_{n/c} = u + n - j \right] - 1_{\{j \leq n\}} \sum_{i=1}^{n-j} \mathbb{P} \left[S_{i/c} = u + i \right] \mathbb{P} \left[S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i}. \end{aligned}$$

Notice that for $u = 0$, this formula corresponds to the formula for V_0 given in (Loisel, Mazza and Rullière(2007)). We can finally give a more compact version of the formula:

THEOREM 5 (Link with the partly shifted process). *Set*

$$\varphi^x(u, t) = \mathbb{P} \left[R_s^x \geq 0 \forall s < t \mid R_0^x = u \right],$$

where $R_s^x = R_s - 1_{\{U < s\}}x$, $s \geq 0$, and U is uniform on $[0, t]$. Then we have

$$V_u = V_Y \left[\lambda t \varphi^Y(u, t) \right].$$

REMARK 3. *Using the notation of (Loisel, Mazza and Rullière(2007)), one can check that V_u is also the variance of*

$$IF_Y \left[\varphi(u, n) \right].$$

This is the variance of

$$\lambda \frac{n}{c} \varphi^Y(u, n),$$

where

$$\varphi^x(u, n) = \mathbb{P} \left[R_s^x \geq 0 \forall s < n/c \mid R_0^x = u \right],$$

and $\varphi(u, n) = \varphi^0(u, n)$ are the probabilities of non ruin before time n/c for the regular and the modified processes.

4. Link with influence functions

The k^{th} -order derivative of the ruin probability with respect to λ involves R_t^x since

$$\frac{\partial^k}{\partial \lambda^k} \varphi(0, t) = t^k \sum_{i=0}^k \binom{k}{i} (-1)^{k-i} \mathbb{E}_{Y_1, \dots, Y_i} \left[\varphi^{Y_1 + \dots + Y_i}(0, t) \right],$$

as a consequence of equation (9) in (Loisel, Mazza and Rullière(2007)). In particular,

$$\frac{\partial}{\partial \lambda} \varphi(0, t) = t \mathbb{E}_Y \left[\varphi^Y(0, t) - \varphi(0, t) \right],$$

as the Y_i are i.i.d., and distributed as generic claim amount W .

4.1. IF FOR THE COMPOUND POISSON DISTRIBUTION

We use here some basic facts given in (Loisel, Mazza and Rullière(2007)). Let \mathbf{T} be a functional of the distribution function F . By definition of the influence function, one has

$$\text{IF}_x[\mathbf{T}] = \lim_{s \downarrow 0} \frac{\mathbf{T}(F^{(s,x)}) - \mathbf{T}(F)}{s},$$

where $F^{(s,x)}$ is defined for $x \in \mathbf{R}$ and $s > 0$ as

$$\text{for } u \in \mathbf{R}, \quad F^{(s,x)}(u) = s 1_{\{x \leq u\}} + (1-s)F(u).$$

THEOREM 6 (IF of S_t). *For any Borel set $A \subset \mathbf{R}$,*

$$\text{IF}_x[\mathbb{P}[S_t \in A]] = \lambda t (\mathbb{P}[x + S_t \in A] - \mathbb{E}_Y[\mathbb{P}[Y + S_t \in A]]),$$

where Y is distributed as W .

Proof: Let n be the number of claims before time t , and set $p_n = \mathbb{P}[N_t = n]$. Then

$$\mathbb{P}[S_t^{(s,x)} \in A] = \mathbb{P}\left[\sum_{n=0}^{+\infty} p_n \mathbb{P}[W^{(s,x)*n} \in A]\right].$$

But note that

$$\mathbb{P}[W^{(s,x)*n} \in A] = \sum_{k=0}^n \binom{n}{k} s^k (1-s)^{n-k} \mathbb{P}[kx + W^{*n-k} \in A].$$

Taking the derivative of the above expression with respect to s when $s = 0$, one obtains

$$\text{IF}_x \mathbb{P}[S_t \in A] = \sum_{n=0}^{\infty} \left(p_n \binom{n}{1} \mathbb{P}[x + W^{*n-1} \in A] - p_n \binom{n}{0} n \mathbb{P}[W^{*n} \in A] \right),$$

and it follows that, using $p_n = \lambda t p_{n-1}/n$ when $n \geq 1$,

$$\text{IF}_x \mathbb{P}[S_t \in A] = \lambda t \sum_{n=0}^{\infty} (p_n \mathbb{P}[x + W^{*n} \in A] - p_n \mathbb{P}[Y + W^{*n} \in A]),$$

where Y is distributed as W . □

4.2. IF ASSOCIATED WITH RUIN PROBABILITIES

THEOREM 7 (IF associated with $\varphi(0, t)$). *The influence function of the ruin probability starting from $R_0 = 0$ is given by*

$$\text{IF}_x[\varphi(0, t)] = \lambda t \left(\varphi^x(0, t) - \text{E}_Y \left[\varphi^Y(0, t) \right] \right),$$

where $\varphi^x(0, t)$ is the ruin probability of the modified risk process R_t^x defined previously, where $\varphi^x(0, t) = \frac{1}{ct} \text{E}[(ct - x - S_t)_+]$.

Proof: Set $\text{IF}_{x,j}(n) = \text{IF}_x \text{P} \left[S_{n/c} = j \right]$, and consider

$$\text{IF}_x \left[\varphi \left(0, \frac{n}{c} \right) \right] = \sum_{j=0}^n \frac{n-j}{n} \text{IF}_{x,j}(n),$$

which can be obtained by using Takac's results (see (Loisel, Mazza and Rullière(2007)), Proposition 9). The result follows then directly from the Definition of $\varphi^x(0, n/c)$ and the interpretation of $\varphi^x(0, n/c)$ in the first Sections. \square

REMARK 4. *Theorem 7 links the derivative with respect to λ of the ruin probability with the influence function since, as we already checked,*

$$\frac{\partial}{\partial \lambda} \varphi(0, t) = t \text{E}_Y \left[\varphi^Y(0, t) - \varphi(0, t) \right].$$

THEOREM 8 (IF associated with $\varphi(u, t)$). *We now turn to the computation of the influence function associated with the ruin probability starting from the initial reserve u , which is given by*

$$\text{IF}_x[\varphi(u, t)] = \lambda t \left(\varphi^x(u, t) - \text{E}_Y \left[\varphi^Y(u, t) \right] \right),$$

where $\varphi^x(u, t)$ is the probability of ruin associated with the modified risk process R_t^x , when $R_0^x = u$.

Proof: We use Proposition 10 of (Loisel, Mazza and Rullière(2007)):

$$\begin{aligned} \text{IF}_x \left[\varphi \left(u, \frac{n}{c} \right) \right] &= \sum_{i=0}^{u+n} \text{IF}_{x,i}(n) - \sum_{k=1}^n \text{IF}_{x,u+k}(k) \varphi(0, \frac{n-k}{c}) \\ &\quad - \sum_{k=1}^n h_{u+k}(k) \text{IF}_x \left[\varphi(0, \frac{n-k}{c}) \right], \end{aligned}$$

From the two previous Theorems, one can transform influence functions as function of the distribution of $x + S_t$ and $Y + S_t$. One then gets the result by using that

$$\begin{aligned} \text{P} \left[T_u^x > \frac{n}{c} \right] &= \text{P} \left[x + S_{n/c} < u + n \right] \\ &\quad - \sum_{i=1}^{n-1} \frac{i}{n} \text{P} \left[x + S_{i/c} = u + i \right] \sum_{j=1}^{n-i} \text{P} \left[S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i} \\ &\quad - \sum_{i=1}^{n-1} \frac{n-i}{n} \text{P} \left[S_{i/c} = u + i \right] \sum_{j=1}^{n-i} \text{P} \left[x + S_{(n-i)/c} = j \right] \frac{n-i-j}{n-i}. \end{aligned}$$

\square

REMARK 5. *Note that one gets back logical expressions for the asymptotic variances:*

$$\begin{aligned} V_0 &= V_Y \left[\lambda t \varphi^Y(0, t) \right] = V_Y \left[IF_Y [\varphi(0, t)] \right], \\ V_u &= V_Y \left[\lambda t \varphi^Y(u, t) \right] = V_Y \left[IF_Y [\varphi(u, t)] \right], \quad u > 0, \end{aligned}$$

as well as the following identities:

$$\begin{aligned} E_Y [IF_Y [\varphi(0, t)]] &= 0, \\ E_Y [IF_Y [\varphi(u, t)]] &= 0, \quad u > 0. \end{aligned}$$

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